RUN LENGTH BASED CONNECTED COMPONENTS AND CONTOUR FOLLOWING FOR ENHANCING THE PERFORMANCE OF CIRCLED REGION EXTRACTION ALGORITHM

BACKGROUND AND SUMMARY OF THE INVENTION

The present invention relates generally to image processing. More particularly, the invention relates to a method and system for identifying contours within pixel-based image data. The identified contours may be used, for example, to identify circled regions on a scanned document, allowing the image processing system to extract content within the circled region to index the document or perform other processing tasks based on the information.

There is considerable interest today in intelligent, automated document imaging systems that can store and retrieve scanned documents and other pixel-based image data with minimal document coding by the user. In a conventional document imaging system the scanned image is displayed on the computer screen, and the user may highlight a portion of the scanned image from which indexing information may be extracted. For example, the user could highlight the caption of a scanned newspaper article, and the caption is processed through Optical Character Recognition (OCR) and then used as an index or label for that scanned image. The index or label is used to later retrieve the scanned newspaper clipping.

While the use of the computer display to highlight relevant portions of a document works well in some applications, there are other applications where physical interaction with the computer screen is not convenient. In such instances, it may be more convenient for the user to simply encircle the region of

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interest on the hard copy document prior to scanning it into the system. In other words, the user would use a pencil or pen to draw a circle around the region of interest, and the image processing system would then identify the user's hand-drawn circle, extract and process the text within the circled region, and use it for indexing or document processing purposes.

Identifying a circled region within the pixel-based image data of a scanned document is a fairly challenging problem. Traditional algorithms for identifying circled regions attempt to identify connected components that have a profile of attributes indicating that they might constitute a circled region. Traditionally, a connected component comprises a collection of black, contiguously adjoining pixels. Depending on the algorithm used, adjoining pixels are those that lie in the immediate neighborhood surrounding a given pixel, typically either the four pixels arranged like points on a compass, or the eight pixels including those four plus diagonals. When a conventional connected component analysis is performed, the scanned image may contain a number of connected component candidates which could represent a user-drawn circle. Post processing algorithms are performed on the connected component data to rule out those that are too small, and thus more likely to correspond to individual letters or noise data. Post processing algorithms also examine the connected component data to determine if there is closure within a given connected component. In this way, closed circles are identified and other markings such as marginal lines or underlining are eliminated.

While conventional connected component analysis does a reasonably good job of identifying potential circled region candidates, the process is fairly

computationally expensive. The present invention addresses this issue through a far more efficient algorithm. Instead of seeking connected components in the conventional fashion--a process that will potentially generate many candidates that each must be analyzed--the present algorithm uses a single data structure

5   that is populated with run length information as the image is being scanned. As will be more fully explained herein, the data structure maintains parent, child and sibling relationship information about each "run" of contiguous pixels within a given scanned row. The parent, child and sibling information is then traversed to identify contours that represent closed circles of suitable size. The run length

10   algorithm thus eliminates the need to painstakingly explore each pixel to its neighboring pixels in the conventional connected component fashion. Moreover, although the algorithm is well suited for identifying circled regions, in its more general case, the algorithm can be used to identify a variety of different contours that meet a set of predefined criteria. Also, while the present "black-and-white"

15   embodiment seeks to identify runs of contiguous black pixels, the algorithm can readily be extended to accommodate gray scale and color pixel image data as well. In such applications, the system merely needs to identify which pixel states or pixel values (e.g., shades or colors) may constitute a member of a run, and the algorithm will identify contours made up of pixels having any of the identified

20   states or values.

According to one aspect of the invention, the method for identifying contours within pixel-based image data comprises expressing the image data as a grid of columns and rows. A scan order is then established over the grid to define parent-child relationships between contiguous pixels in adjacent rows and

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sibling relationships among non-contiguous pixels in the same row. A run data structure is established in the computer-readable memory that defines a run member by its row position and by its starting and ending column positions. The run data structure further defines parent, child and sibling structures for storing information about these relationships.

As the image data is scanned according to the scan order established above, the processor identifies contiguous pixels of a predetermined state as identified run members. It also determines the parent-child and sibling relationships of the identified run members. This information is populated into the run data structure with the row position and starting and ending column positions of the identified run member and with the parent-child and sibling relationships of the identified run member. The populated data structure is then traversed by following the parent-child and sibling relationships to identify contours within the pixel-based image data. If desired, once the contours have been identified, additional processing can be performed to improve the likelihood that the identified contour represents a user-drawn circle. The additional processing aids in discriminating one user-drawn circle from another, when two circles overlap. The additional processing involves computing feature points along the contour and then breaking the contour into segments at the feature points. The broken contours are then used to generate reconstructed circles, each having its own identity and thus each being separate from other circles which it may overlap.

For a more complete understanding of the invention, its objects and advantages, refer to the following specification and to the accompanying drawings.

BRIEF DESCRIPTION OF THE DRAWINGS

Figure 1 is a pixel diagram illustrating the concepts of run length, parent-child relationships and sibling relationships;

Figure 2 is a data structure diagram illustrating a presently preferred data structure for implementing the invention;

Figure 3 is a flowchart diagram illustrating a presently preferred algorithm for populating the data structure of Figure 2;

Figure 4 is a pixel diagram illustrating an example of a contour, useful in understanding how the data structure of Figure 2 is populated by applying the algorithm of Figure 3;

Figure 5 is a flowchart diagram illustrating how the data structure is traversed to identify a contour;

Figure 6 is a pixel diagram illustrating the contour following process of Figure 5 as applied to the exemplary data of Figure 4;

Figure 7 is a pixel diagram illustrating the run length midpoints with darker shading;

Figure 8 is a pixel diagram, similar to Figure 6, but showing the two main contours with different line styles to illustrate how the contours are joined; and

Figure 9 is a flowchart diagram of an enhanced circled region extraction algorithm suitable for discriminating among overlapping circles.

DESCRIPTION OF THE PREFERRED EMBODIMENT

The run length based method of the preferred embodiment scans image data in a pre-established scan order to identify parent, child and sibling structures within the image data. In this regard, the image data may be expressed as a rectangular grid **10** of columns and rows comprising individual pixels that are represented in Figure 1 as squares **12**. In the general case, each pixel may contain a data value representing a particular state, tone or color of which the overall image is made. In black-and-white images, each pixel contains binary data indicating whether that pixel is black or white. In a gray scale image, each pixel may contain a byte or word of data indicating what gray scale tone the pixel represents. In a color image the pixel may contain data indicating the color and intensity of the pixel. The present invention works with all forms of image data, regardless of whether the image is black and white, gray scale or color.

As will be more fully explained below, the processing algorithm scans the grid of pixel data in a predetermined scan order. For illustration purposes, it will be assumed that the scan order is from top to bottom and from left to right. Thus the grid **10** would be scanned beginning at pixel **14** in the upper left-hand corner and ending with pixel **16** in the lower right-hand corner.

The algorithm is designed to identify and correlate groups of adjacent pixels that define a linear string termed a *run* or *run length*. The scanning process groups pixels of a predetermined state together to form run lengths within the image.

In Figure 1 a black-and-white image is illustrated. White pixels appear as white rectangles, such as pixels **12**. Black pixels appear as black dots within the pixel square, such as pixels **18**. In the normal case, the system is designed to form run lengths from contiguous opaque pixels, such as the black pixels in

5 Figure 1. However, in the more general case, the system can be used to form run lengths of pixels occupying a variety of different states. For example, if gray scale images are provided, the system could be configured to consider all pixels darker than a predetermined gray scale level to be considered in forming run lengths. In a color image, colors of predetermined intensities or tonal hues may

10 be selected for potential membership in a run length.

The preferred embodiment considers run lengths to be defined by a collection of contiguous pixels of a predetermined state (e.g., the black pixel state in a black-and-white image). Three such run lengths are illustrated in Figure 1 at **20**, **22** and **24**. In Figure 1 bounding box rectangles have been

15 shown around each of these run lengths to make them more visible. Each run length may be represented by its x and y coordinates, that is its row and column coordinates within grid **10**. Because the scan order is performed on a row by row basis, each run length can be identified by its beginning and ending column positions and its row position. These three coordinate positions that uniquely

20 identify a run length are stored as the x-minimum **36**, x-maximum **38** and y-coordinate **40** data values within the run data structure **30** shown in Figure 2. Specifically, the run data structure **30** comprises a linked list that includes data values for storing the x-minimum, x-maximum and y-coordinate values for each run length identified during the scanning process.

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Individual pixels (such as the black pixels in Figure 1) are identified as a run length if they occupy contiguous pixel locations within a given row. To accommodate slight imperfections and data dropout in the image, the algorithm can be configured to ignore single pixels (or more) of the incorrect state within a

5   potential run length. This may be done through preprocessing or on the fly as the scanning algorithm performs its other tasks. In effect, if a single white pixel is encountered in what would otherwise be a complete run length, the algorithm can treat that pixel as if it were a black pixel, thereby assigning it to the run length, provided the white pixel is neighbored on both sides by black pixels.

10   The scanning algorithm and the associated data structure of Figure 2 defines a hierarchy among run lengths. The hierarchy is termed a parent-child-sibling hierarchy. The concept is illustrated in Figure 1. A parent-child relationship exists where two run lengths have one or more adjacent black pixels (or pixels of whatever predetermined state defines a run length) in a given

15   column. In Figure 1, run lengths **20** and **22** have two pairs of vertically adjacent pixels (shown circled and identified by reference numeral **32**) that qualify these run lengths for the parent-child relationship. Specifically, run length **20** is deemed the parent and run length **22** is deemed the child, because run length **20** is scanned before run length **22**. The existence of this parent-child

20   relationship is discovered automatically as the scanning algorithm performs its tasks, and the relationship is recorded in the run data structure **30**. The details of how this is accomplished will be discussed below in connection with the flowchart of Figure 3.

The scanning algorithm and its associated data structure **30** also defines a sibling relationship between two run lengths that lie on the same row and share the same parent , such as run lengths **22** and **24**. The scanning algorithm automatically detects sibling relationships and stores them in data structure **30**, as will be more fully described in connection with Figure 3.

The presently preferred run data structure **30** defines each run length **34** by its associated x-minimum value **36**, x-maximum value **38** and y-coordinate value **40**. The run length data structure also has associated with it a parent pointer **42**, a child pointer **44** and a sibling pointer **46**. The parent, child and sibling pointers **42-46** establish a linked list of the run length data structures. The scanning algorithm populates this data structure as it visits each pixel in grid **10**. In the preferred embodiment a single, multiply linked list is populated with individual run length data structures **34** corresponding to each run length identified by the scanning algorithm.

Run Length Scanning Algorithm

Figure 3 shows the presently preferred run length scanning algorithm. For purposes of understanding the algorithm it will be assumed that scanning has proceeded to the point where a run length 'k' has been identified and the algorithm is currently processing a run length 'j' that has a sibling run length 's'. The relationship of scan lengths k, j and s are shown diagrammatically at **50** for convenience.

The scanning algorithm begins at step **60** by scanning the image in the predetermined scan order, searching for occurrences of run lengths. As discussed above, run lengths are identified as contiguous pixels in a common

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row that have the same pixel state (e.g. contiguous black pixels). The preferred scanning algorithm is recursive. For each run length identified (such as run length 'j') the following process is performed.

If, as illustrated at step **62**, the identified run length (j) is vertically adjacent a run length (k) from a previous row, then the current run length's parent data structure (j's parent data structure) is populated with a reference to the vertically adjacent run length (k), as shown at **64**. The algorithm then tests k's child data structure at step **66** to determine if it is currently empty (containing a null value). If a null value is found, the algorithm populates k's child data structure with a reference to scan length j, as illustrated at **68**. If k's child data structure is not currently empty (in other words if it contains a reference to another child) the algorithm detects this at step **70** and then populates j's sibling data structure with the value stored as k's child (s), as shown at **72**. Thus steps **66-72** essentially test whether the child of k is null. If so, it sets j to the child of k. Otherwise it sets j as the sibling of the child of k.

In some instances the algorithm will detect that the child already has a sibling. In this case the algorithm sets the child as the sibling's sibling, and so on. This is shown in steps **74** and **76**. When siblings are identified the algorithm sets their respective sibling pointers to each other as shown at step **72** and **78**.

The scanning algorithm illustrated in Figure 3 populates the run data structure for later retrieval of connected components. The connected component retrieval process operates by tracing from parent to child and from sibling to sibling. Figure 4 provides an example of an image of a generally u-shaped contour made up of a plurality of individual run lengths. The objective of the

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connected component retrieval process is to identify this u-shaped contour by tracing the parent-to-child and sibling-to-sibling relationships of the run lengths involved.

Note that the u-shaped contour in Figure 4 has a left side leg **50** comprised of parent-child relationships when scanned from left to right and from top to bottom. The right side leg **52** is also comprised of parent-child relationships. The left and right side legs are joined by the bottom leg **54**. Because the scan order is from left to right and from top to bottom, the point at which bottom leg **54** joins right side leg **52** involves a sibling relationship. Specifically, run length **56** and run length **58** overlap at **60**. Run length **56** is thus treated as the sibling of the parent **62** of run length **58**.

<u>Connected Component Retrieval</u>

Once the run data structure **30** has been populated by the scanning process, a connected component retrieval process is commenced. Referring to the flowchart of Figure 5, the retrieval process makes use of the parent, child and sibling information obtained and stored during initial scanning. This initial screening step is depicted at step **102**.

The connected components that remain are processed as follows. Beginning with the run length occupying the upper left-most corner (determined by its x-minimum, x-maximum and y-coordinate values), the process first identifies, by examining the run data structure, if that run length has a child. This is depicted at step **104** and also illustrated in the connected component retrieval diagram of Figure 6. In Figure 6 the starting point is indicated at **80**. If no child is found, such as in the case of the run length at **82** in Figure 6, the procedure links

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to the sibling, as indicated at **84**. The sibling linking step is depicted at **106** in Figure 5.

The procedure is recursive. It continues to seek child and sibling connections until no more children or siblings are identified, as indicated at step **108**. This condition occurs at **86** in Figure 6. Once no further children or siblings can be found, the procedure begins looking for parents of a given run length as depicted at step **110**. For example, referring to Figure 5, the procedure would link run link **88** with its parent **90**. Eventually, the entire connected component is retrieved, resulting in a linked structure identifying which run lengths are connected to which other run lengths. The entire linked structure thus represents one connected component. Note that it is possible for a given image to have several connected components that are detached from one another. The procedure described in Figure 5 may be repeated multiple times until all connected components have been identified.

While the procedure described so far is well adapted at identifying connected components found in pixel-based images, some image processing tasks, such as identifying circled regions, benefit by further processing. Specifically, the presently preferred embodiment converts the connected component into a contour, which may then be used to generate new structures that are more easily and reliably processed. In an application where user-drawn circles are identified so that the encircled region can be extracted, the connected components are converted into contours that are then used to re-generate circles that are more easily processed by the computer.

The contour following algorithm uses the same data structure that generated the connected components, namely the run data structure **30**. When retrieving connected components, the algorithm recursively retrieves the child or parent of each run length. If no child or parent can be found, the algorithm

5 switches to siblings. For each sibling, the algorithm continues to search for children and parents first. The parent and child relationship ensures that the related run lengths retrieved have pixels that belong to the same contour and are next to each other in a section of the contour.

In the presently preferred embodiment the contour following algorithm

10 defines the mid-point of each run length. The mid-point P is thus defined by its P_X and P_Y components as follows:

$$P_x = \frac{x_{min} + x_{max}}{2}$$

15 $$P_y = y_{coordinate}$$

Figure 7 shows the mid-points of an example run length graph, where the mid-points are marked as dark points and the remaining pixels are marked in lighter cross hatched shading. Note that every time the retrieval switches to one

20 of the siblings, the algorithm starts another contour segment. Thus in Figure 7, there are two contour segments, a left-most contour segment **98** and a right-most contour segment **99**. The two segments are joined along the horizontal run length **96**. Thus they share one mid-point in common, that being mid-point **120**.

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The contour following algorithm assigns the mid-point at the end of a contour to be the beginning mid-point of the subsequent contour. In this way, both contours are linked together, as illustrated in Figure 8. In Figure 8 the contour **98** is shown linked together using solid lines and the contour **99** is shown linked with dotted lines.

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To more fully understand how the run length based connected component analysis may be used in an application, refer to the flow diagram of Figure 9. Figure 9 illustrates how a user-drawn circle may be identified in a scanned image. The user-drawn circle may, for example, encircle a file designation label or number used to index the scanned image for storage and retrieval.

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Beginning at step **200**, the procedure first pre-processes the image to convert the image data into suitable resolution and bit-depth for connected component analysis. This is an optional step that may be performed, for example, to convert color image data into black-and-white data or to convert gray scale data into black-and-white data.

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The procedure next proceeds to step **202** where connected component analysis is performed using the run length analysis techniques described above. If desired, at step **204**, half-tone data may be eliminated and small gaps in pixel data (attributable to noise) may be eliminated. Individual white pixels in an otherwise black pixel domain may be converted to black pixels to "fill in" or de-speckle the image data.

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After the connected components have been identified and the run data structure populated, the process advances to step **206**, where contour segment

information is extracted from the data. The contour segment extraction process is performed as described above, resulting in one or more connected contours, such as the connected contour shown in Figure 8. The connected contour serves as a "replacement" for the originally drawn circle. Feature points are identified in the contour, step **208**, and these feature points are used to break the contour into contour segments, step **210**. Breaking the contour according to its feature points is useful in situations where user-drawn circles may contact or overlap other contour structures. The feature points are then used at step **212** to reconstruct circles, which are then output at step **214**. The reconstructed circles are generated without gaps or ambiguity regarding closure. Thus the subsequent image processing can be performed more efficiently.